

MEME15203 Statistical Inference Marking Guide**Assignment 3****UNIVERSITI TUNKU ABDUL RAHMAN**

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| Faculty: | FES | Unit Code: | MEME15203 |
| Course: | MAC | Unit Title: | Statistical Inference |
| Year: | 1,2 | Lecturer: | Dr Yong Chin Khian |
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Q1. Suppose that X_1, \dots, X_n is a random sample from a Poisson distribution, $X_i \sim \text{POI}(\theta)$,

- (a) Find a complete and sufficient statistic for θ .
- (b) Find the UMVUE for θ
- (c) Find the UMVUE of $e^{-7\theta}$ using Lehmann Scheffe Theorem.
- (d) Find the UMVUE of $e^{-7\theta}$ using Rao Blackwell Theorem.

(20 marks)

Ans.

$$(a) f(x) = e^{-\theta} (1/x!) e^{x \ln \theta} = c(\theta) h(x) e^{q(\theta)t(x)}$$

where $c(\theta) = e^{-\theta}$, $h(x) = 1/(x!)$, $q(\theta) = \ln \theta$, and $t(x) = x$.

Thus the p.d.f. of X belongs to the regular exponential family. Since the p.d.f. of X belongs to the regular exponential family, thus by the theorem, $S = \sum_{i=1}^n X_i$ is a c.s.s of θ

(b) Since $E(\sum_{i=1}^n X_i/n) = E(\bar{X}) = \theta$, thus, \bar{X} is the UMVUE of θ

(c) Let $E(a^S) = e^{-7\theta}$

$$\begin{aligned} E(a^S) &= E(e^{S \ln a}) \\ &= e^{n\theta(e^{\ln a} - 1)} \\ &= e^{n\theta(a - 1)} \\ \implies n(a - 1) &= -7 \\ a &= \frac{n-7}{n} \end{aligned}$$

$$\text{Thus } E\left(\frac{n-7}{n}\right)^S = e^{-7\theta}$$

Since $\left(\frac{n-7}{n}\right)^S$ is a function of the c.s.s. of θ which is an UE of $e^{-7\theta}$, thus $\left(\frac{n-7}{n}\right)^S$ is the UMVUE of $e^{-7\theta}$.

(d) Let

$$T = \begin{cases} 1, & X_1 + \dots + X_7 = 0 \\ 0, & \text{otherwise} \end{cases}.$$

$E(T) = P(X_1 + \dots + X_7 = 0) = e^{-7\theta}$. Thus T is an unbiased estimator of $e^{-7\theta}$. Since S is CSS of θ . Hence by Rao-Blackwell theorem, $T^* = E(T|S)$

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$$\begin{aligned}
& \text{is an UMVUE of } e^{-7\theta}. \\
& E \left[T \mid \sum_{i=0}^n X_i = s \right] \\
&= 1 \cdot P[X_1 + \dots + X_7 = 0 \mid X_1 + X_2 + \dots + X_n = s] \\
&= \frac{P(X_1 + \dots + X_7 = 0, X_8 + \dots + X_n = s)}{P(X_1 + \dots + X_n = s)} \\
&= \frac{P(X_1 + \dots + X_n = s)}{P(X_1 + \dots + X_7 = 0) \times P(X_8 + \dots + X_n = s)} \text{ Since } X_1, X_2, \dots, X_n \text{ are independent.} \\
&= \frac{e^{-7\theta} [(n-7)\theta]^s e^{-(n-7)\theta}}{(n\theta)^s s! e^{-n\theta} / s!} \\
&= \left(\frac{n-7}{n} \right)^s \\
&= \left(\frac{n-7}{n} \right)^s
\end{aligned}$$

Q2. Let X_1, X_2, \dots, X_n be random sample of size n from an Exponential distribution with unknown mean θ . Find the UMVUE of $\gamma = e^{-t/\theta}$ using Rao-Blackwell theorem.

(20 marks)

Ans.

Let

$$T = \begin{cases} 1, & X_1 > t \\ 0, & \text{otherwise} \end{cases}.$$

Then, $E(T) = P(X > t) = e^{-t/\theta}$. Thus T is an unbiased estimator of γ .
 $f(x) = \frac{1}{\theta} e^{-x/\theta} = c(\theta)h(x)e^{q(\theta)t(x)}$, where $c(\theta) = \frac{1}{\theta}$, $h(x) = 1$, $q(\theta) = \frac{1}{\theta}$, and $t(x) = x$, hence $f(x)$ is a member of $REC(\theta)$ and $S = \sum_{i=1}^n X_i$ is a complete sufficient statistics for θ .

Thus, by Rao-Blackwell theorem, $T^* = E(T|S)$ is an UMVUE of γ .

$$\begin{aligned}
f_{X_1, S}(x_1, s) &= f_{X_1, S_1}(x_1, s - x_1) \text{ where } S_1 = X_2 + \dots, X_n \sim \text{gamma}(n-1, \theta) \\
&= f_{X_1}(x_1) f_{S_1}(s - x_1) \\
&= \frac{1}{\theta} e^{-x_1/\theta} \frac{1}{\Gamma(n-1)\theta^{n-1}} (s - x_1)^{n-2} e^{-(s-x_1)/\theta} \\
&= \frac{1}{\Gamma(n-1)\theta^{n-1}} (s - x_1)^{n-2} e^{-s/\theta}
\end{aligned}$$

$$f_{X_1|S}(x_1) = k(s - x_1)^{n-2}, 0 < x_1 < s$$

$$\int_0^s k(s - x_1)^{n-2} dx_1 = 1$$

$$k \left[\frac{(s-x_1)^{n-1}}{-(n-1)} \right]_0^s = 1$$

$$k \left[\frac{s^{n-1}}{n-1} \right] = 1$$

$$k = \frac{n-1}{s^{n-1}}$$

$$\therefore f_{X_1|S}(x_1) = \frac{(n-1)(s-x_1)^{n-2}}{s^{n-1}}, 0 < x_1 < s$$

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$$\begin{aligned}
E[T|S] &= P[X_1 > t|s] \\
&= \int_t^s \frac{(n-1)(s-x_1)^{n-2}}{s^{n-1}} dx_1 \\
&= \frac{n-1}{s^{n-1}} \left[\frac{(s-x_1)^{n-1}}{-(n-1)} \right]_t^s \\
&= \left(\frac{n-1}{s^{n-1}} \right) \left(\frac{(s-t)^{n-1}}{n-1} \right) \\
&= \left(\frac{s-t}{s} \right)^{n-1}
\end{aligned}$$

Q3. Let X_1, X_2, \dots, X_n be a random sample from a distribution with pdf

$$f(x; \theta) = 8\theta x^{8\theta-1} I_{(0,1)}(x).$$

Find the UMVUE of θ ,

(20 marks)

Ans.

$f(x; \theta) = 8\theta x^{8\theta-1} = 8\theta e^{(8\theta-1)\ln x} = c(\theta)h(x)e^{q(\theta)t(x)}$ which is a member of $\text{REC}(\theta)$ with $q(\theta) = 8\theta - 1$ and $t(x) = \ln x$. Thus, $T = \sum \ln X_i$ is a complete sufficient statistic for θ .

Let $v_i = -\ln(x_i)$. Thus $0 < v_i < \infty$. This correspond to a 1-1 transformation of $x_i = e^{-v_i}$

$$h^{-1}(v_i) = e^{-v_i}$$

$$f_V(v_i) = f_X(h^{-1}(v_i)) \left| \frac{dh^{-1}(v_i)}{dv_i} \right| = 8\theta e^{-(8\theta-1)v_i} e^{-v_i} = 8\theta e^{-8\theta v_i}$$

$$\Rightarrow V_i \sim \text{EXP}(1/8\theta) \text{ and}$$

$$U = -\sum_{i=1}^n \ln x_i = \sum_{i=1}^n V_i \sim \text{gamma}(\alpha = n, \beta = \frac{1}{8\theta})$$

$$\begin{aligned}
E(U^{-1}) &= \int_0^\infty u^{-1} \frac{(8\theta)^n}{\Gamma(n)} u^{n-1} e^{-8\theta u} du \\
&= \frac{(8\theta)^n}{\Gamma(n)} \int_0^\infty u^{n-2} e^{-8\theta u} du \\
&= \frac{(8\theta)^n}{\Gamma(n)} \left[\frac{\Gamma(n-1)}{(8\theta)^{n-1}} \right] \\
&= \frac{8\theta}{n-1}
\end{aligned}$$

$$E\left(\frac{n-1}{8} U^{-1}\right) = \theta$$

Thus by Lehmann Scheffie Theorem, $\frac{n-1}{8U}$ is an UMVUE of θ

Q4. Let X_1, X_2, \dots, X_n be random sample of size n from $f(x|\theta) = \binom{k}{x} \theta^x (1-\theta)^{k-x}$, $x = 0, 1, \dots, k$. Find the uniformly minimum variance unbiased estimator (UMVUE) of $g(\theta) = \binom{k}{2} \theta^2 (1-\theta)^{k-2}$.

(20 marks)

MEME15203 Statistical Inference Marking Guide*Ans.*

$f(x|\theta) = \binom{k}{x} \theta^x (1-\theta)^{k-x} = \binom{k}{x} (1-\theta)^k e^{x \ln[\theta(1-\theta)]}$ which is $REC(q_1)$ with $q_1(\theta) = \ln[\theta(1-\theta)]$ and $t_1(x) = x$. By the theorem, $S = \sum_{i=1}^n X_i$ is a complete sufficient statistic for θ .

Let

$$T = \begin{cases} 1, & X_1 = 2 \\ 0, & \text{otherwise} \end{cases}.$$

$E(T) = P(X = 2) = \binom{k}{2} \theta^2 (1-\theta)^{k-2} = g(\theta)$. Thus T is an unbiased estimator of θ . Since S is CSS of θ . Hence by Rao-Blackwell theorem, $T^* = E(T|S)$ is an UMVUE of $g(\theta)$.

$$\begin{aligned} E[T | \sum_{i=1}^n X_i = s] &= 1 \cdot P[X_1 = 2 | X_1 + X_2 + \dots + X_n = s] \\ &= \frac{P(X_1=2, X_2+\dots+X_n=s-2)}{P(X_1+\dots+X_n=s)} \\ &= \frac{P(X_1=2) \times P(X_2+\dots+X_n=s-2)}{P(X_1+\dots+X_n=s)} \text{ Since } X_1, X_2, \dots, X_n \text{ are independent and} \\ &S \sim \text{Bin}(kn, \theta), X_2 + \dots + X_n \sim \text{Bin}(k(n-1), \theta) \\ &= \frac{\binom{k}{2} \theta^2 (1-\theta)^{k-2} \times \binom{k(n-1)}{s-2} \theta^{s-2} (1-\theta)^{k(n-1)-(s-2)}}{\binom{kn}{s} \theta^s (1-\theta)^{kn-s}} \\ &= \frac{\binom{k}{2} \times \binom{k(n-1)}{s-2}}{\binom{kn}{s}} \end{aligned}$$

Q5. Let X_1, \dots, X_{30} be a random sample from a distribution with probability density function(p.d.f.)

$$f(x) = \frac{\theta^6}{\Gamma(6)} x^5 e^{-\theta x} I(0, \infty), \theta > 0.$$

- (a) Show that the p.d.f. of X belongs to the regular exponential family.
- (b) Find a complete and sufficient statistic for θ .
- (c) Find the UMVUE for $V(X_1)$.
- (d) Find the UMVUE for θ .

(20 marks)

Ans.

$$\begin{aligned} \text{(a)} f(x) &= \frac{\theta^6}{\Gamma(6)} x^5 e^{-\theta x} I(0, \infty), \theta > 0 \\ &= c(\theta) h(x) e^{q(\theta)t(x)} \\ \text{where } c(\theta) &= \theta^6, h(x) = 1/\Gamma(6) x^5, q(\theta) = -\theta, \text{ and } t(x) = x. \\ \text{Thus the p.d.f. of } X &\text{ belongs to the regular exponential family.} \end{aligned}$$

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(b) Since the p.d.f. of X belongs to the regular exponential family, thus by the theorem, $S = \sum_{i=1}^{30} X_i$ is a c.s.s of θ

(c) $X \sim GAM(6, \frac{1}{\theta})$ and $S \sim GAM(180, \frac{1}{\theta})$

Let $g(\theta) = V(X_1) = \frac{6}{\theta^2}$

$$E(S^2) = V(S) + E^2(S) = \frac{180}{\theta^2} + \frac{180^2}{\theta^2} = \frac{32580}{\theta^2}$$

$$\text{Thus } E\left[\frac{S^2}{5430}\right] = \frac{32580}{5430\theta^2} = \frac{6}{\theta^2}$$

Since $E\left[\frac{S^2}{5430}\right] = \frac{6}{\theta^2}$, thus, by Lehmann Scheffe Theorem, $\frac{S^2}{5430}$ is the UMVUE of $V(X_1)$.

$$(d) E(S^{-1}) = \int_0^\infty s^{-1} \frac{\theta^{180}}{\Gamma(180)} s^{179} e^{-\theta s} ds = \int_0^\infty \frac{\theta^{180}}{\Gamma(180)} s^{178} e^{-\theta s} ds = \frac{\theta^{180}}{\Gamma(180)} \frac{\Gamma(179)}{\theta^{179}} = \frac{\theta}{179}$$

Thus $E(179S^{-1}) = \theta$.

Since $E(179S^{-1}) = \theta$, thus, by Lehmann Scheffe Theorem, $179S^{-1}$ is the UMVUE of θ .